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The PAU Survey: spectral features and galaxy clustering using simulated narrow-band photometry

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ABSTRACT

We present a mock catalogue for the Physics of the Accelerating Universe Survey (PAUS) and use it to quantify the competitiveness of narrow-band imaging for measuring spectral features and galaxy clustering. The mock agrees with observed number count and redshift distribution data. We demonstrate the importance of including emission lines in the narrow-band fluxes. We show that PAUCam has sufficient resolution to measure the strength of the 4000 Å break to the nominal PAUS depth. We predict the evolution of a narrow-band luminosity function and show how this can be affected by the O II emission line. We introduce new rest-frame broad-bands (UV and blue) that can be derived directly from the narrow-band fluxes. We use these bands along with D4000 and redshift to define galaxy samples and provide predictions for galaxy clustering measurements. We show that systematic errors in the recovery of the projected clustering due to photometric redshift errors in PAUS are significantly smaller than the expected statistical errors. The galaxy clustering on two halo scales can be recovered quantitatively without correction, and all qualitative trends seen in the one halo term are recovered. In this analysis, mixing between samples reduces the expected contrast between the one halo clustering of red and blue galaxies and demonstrates the importance of a mock catalogue for interpreting galaxy clustering results. The mock catalogue is available on request at <https://cosmohub.pic.es/home>.

Key words: galaxies: evolution – galaxies: formation – galaxies: luminosity function, mass function – large-scale structure of Universe.

1 INTRODUCTION

Clustering measurements at low redshifts have been shown to display a dependence on galaxy properties such as stellar mass, lu-

minosity, and colour, which suggest that these properties depend on the mass of the host dark matter halo (e.g. Norberg et al. 2002; Zehavi et al. 2011). Galaxy clustering measurements are therefore not only useful for constraining the cosmological model but also for developing our understanding of galaxy formation physics.

The processes that shape how the efficiency of galaxy formation depend on halo mass may change with redshift, so it is important to extend measurements of galaxy clustering as a function of intrinsic galaxy properties to higher redshift. One clear piece of evidence hinting at evolution in the galaxy formation process is the dramatic

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change in the amount of star formation activity since $z \sim 1-2$, with roughly 10 times less star formation globally by the present day (Lilly et al. 1996; Madau et al. 1996).

The measurement of clustering as a function of galaxy properties poses different challenges to those faced when using large-scale structure to constrain cosmological parameters. In the cosmological case, the aim is to maximize the volume probed whilst maintaining an appropriate number density of galaxies to achieve a moderate signal-to-noise ratio in the power spectrum measurement (e.g. Feldman, Kaiser & Peacock 1994). The signal-to-noise ratio can be boosted by targeting galaxies with stronger clustering or a larger bias than the average population; beyond this, the selection of the galaxies is not that important in the cosmological case. On the other hand, when using clustering to probe galaxy formation, the desire is for a high number density of galaxies with a uniform selection covering a wide baseline in the intrinsic galaxy property of interest.

Progress towards compiling large-scale structure samples for galaxy formation studies at intermediate redshifts has been made through the Galaxy And Mass Assembly Survey (GAMA; Driver et al. 2011), which targets galaxies in the r -band brighter than $r = 19.8$, with a median redshift of $z \sim 0.2$ over 286 deg^2 with high completeness, and the VIMOS Public Extragalactic Redshift Survey (VIPERS), which obtained redshifts for 867 765 galaxies with $i_{\text{AB}} < 22.5$ over 24 deg^2 at ~ 47 per cent completeness (Scodreggio et al. 2018). The PRIMUS Multi-object Survey (PRIMUS; Coil et al. 2011) used slit masks to measure ~ 2500 redshifts in a single telescope pointing, recording 130 000 redshifts over 9.1 deg^2 to $i_{\text{AB}} = 23.5$, with a redshift distribution peaking at $z \sim 0.6$. These surveys have been used to carry out a large number of analyses to quantify the galaxy populations and to constrain the cosmological model. Next, we highlight some results from these surveys that explicitly focus on using galaxy clustering measurements to probe the physics of galaxy formation. Farrow et al. (2015) measured galaxy clustering as a function of luminosity and colour using GAMA. Loveday et al. (2018) inferred the pairwise velocity distribution using the small-scale galaxy clustering measured from GAMA. In both cases, these observational results were compared to theoretical models of the sort we will use here. Marulli et al. (2013) used VIPERS to measure the dependence of galaxy clustering on stellar mass and luminosity for $0.5 < z < 1.1$. Coupon & Arnouts (2015) combined clustering measurements with a gravitational lensing analysis to constrain the galaxy halo connection. Skibba et al. (2014) measured the clustering of galaxies in PRIMUS as a function of colour and luminosity. Skibba et al. (2015) studied the variation of the clustering amplitude with stellar mass, and Bray et al. (2015) examined how the luminosity dependence of clustering depends on pair separation.

A limitation of spectroscopic surveys is the number of redshifts that can be measured in a single telescope pointing. This is set by the number of fibres or slits available to deploy to measure galaxy redshifts in the field of view. The use of some form of aperture to capture the light from a single galaxy also introduces a systematic effect on the clustering measured on small scales. The physical size of the slit or fibre means that in some cases only one member of a pair of galaxies within a particular angular separation can be targeted for a redshift measurement. This ‘fibre collision’ effect can be mitigated by repeat observations of the same field or by applying a correction to the measured pair counts.

An alternative to using spectroscopy to measure the radial distance to a galaxy is to use photometry taken in a number of bands. A photometric redshift can be assigned to a galaxy by, for example, comparing the observed flux in different bands to that derived from a template spectrum that is shifted in redshift (Benítez 2000; Bol-

zonella, Miralles & Pelló 2000). The photometric redshift approach has three advantages over spectroscopy: (1) the galaxy selection is homogeneous down to the flux limit, without any bias towards a higher success rate of redshift measurement for galaxies with emission lines (although the precision and catastrophic error rate of photometric redshifts will vary for different populations of galaxies; see e.g. Martí et al. 2014; Sánchez et al. 2014), (2) there are no ‘fibre collisions’ that can impact galaxy clustering measurements, and (3) there is no requirement to match the surface density galaxies to the number of slits or fibres within the field of view.

Broad-band photometry, in which the typical filter width is $\sim 1000 \text{ Å}$, is limited to a redshift precision $\Delta z/(1+z)$ (hereafter σ_z) of $\sim 3-5$ per cent. CFHTLS wide, a broad-band survey observing in u, g, r, i , and z that is 80 per cent complete to $i < 24.8$ reaches $\sigma_z \sim 3$ per cent for $i < 24$ with ~ 4 per cent catastrophic errors (defined as $\sigma_z > 15$ per cent; Ilbert 2012). This level of precision is sufficient to divide galaxies into redshift shells in which the projected clustering can be measured. The error in the radial distance estimate in this case is $\sim 100 h^{-1} \text{ Mpc}$ at $z = 0.7$.

The accuracy of photometric redshifts can be improved using narrower filters (Wolf et al. 2004). The Advanced Large, Homogeneous Area Medium Band Redshift Astronomical Survey (Moles et al. 2008) offers a recent example of this using 20 medium band filters, each $\sim 300 \text{ Å}$ in width, to reach an accuracy of $\sigma_z = 1.4$ per cent for galaxies with $i < 24.5$ (Molino et al. 2014). Ilbert et al. (2009) reached $\sigma_z = 1.2$ per cent for objects with $i < 24$ over the 2 deg^2 COSMOS field using a combination of broad, medium, and narrow bands spanning the ultraviolet (UV) to the mid-infrared.

The Physics of the Accelerating Universe Survey (PAUS) is a narrow-band imaging survey using PAUCam (Padilla et al. 2016), which was commissioned in 2015 June, on the 4.2 m William Herschel Telescope. PAUS measures narrow-band fluxes using forced photometry on objects previously detected in overlapping broad-band photometric surveys CFHTLenS (Heymans et al. 2012) and KiDS (Kuijken et al. 2015). PAUS aims to perform forced photometry measurements in 40 narrow bands over 100 deg^2 for objects $i < 23$, and reach signal-to-noise of 3 at narrow-band magnitude 23. Each of the 40 narrow-band filters has FWHM 130 Å and are spaced by 100 Å , over the wavelength range of $4500-8500 \text{ Å}$ (Martí et al. 2014). Fig. 1 shows the PAUCam narrow-band filters compared to the g, r , and i bands from CFHTLS. 40 narrow bands span the region covered by these three broad-band filters. The increased spectral resolution of PAUS imaging will allow for photometric redshift measurements of $\sigma_z = 0.35$ per cent for objects $i < 23$ (Martí et al. 2014). This represents an improvement of nearly an order of magnitude compared with typical broad-band redshift measurement uncertainties, and in principle allows the radial distance information to be used in clustering estimates and to infer membership of galaxy groups. The PAUS collaboration has recently started to release early scientific results. Tortorelli et al. (2018) built PAUS image simulations using a forward modelling approach, and Cabayol et al. (2018) used the PAUS 40 narrow bands for star-galaxy separation.

The spectral features of a galaxy encode information about intrinsic properties such as its stellar mass, age, and metallicity. Using these properties to define samples for clustering studies can then help us to understand the connection between galaxy properties and the mass of the host dark matter halo. These features include emission lines, absorption features, the 4000 Å break, and the shape of the continuum. Measuring the spectral features of individual galaxies has largely been in the domain of spectroscopic surveys. Kauffmann et al. (2003) used a combination of the strength of the 4000 Å break and the $H\delta$ absorption feature to constrain the stel-

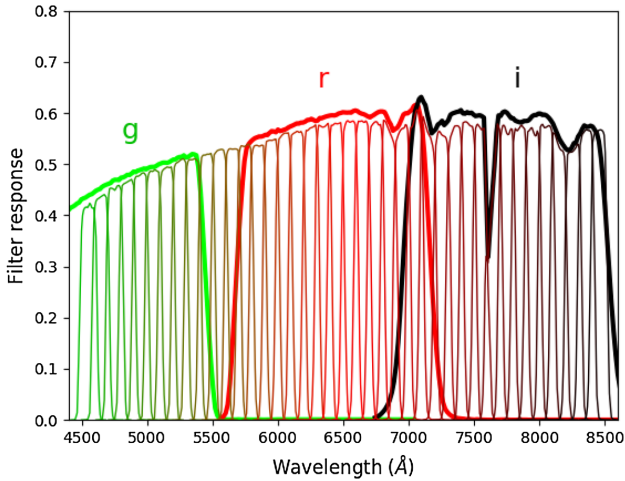


Figure 1. Filter response as a function of wavelength for the 40 PAUcam filters (thin lines) compared to CFHT MegaCam broad-band filters *g*, *r*, and *i* (thick lines). Filter response curves include atmospheric transmission, telescope optics, and CCD quantum efficiency.

lar age, and contribution to stellar mass from recent star formation events, for a large sample of galaxies drawn from the spectroscopic Sloan Digital Sky Survey (SDSS). Kriek et al. (2011) used stacking to measure the average values of spectral features using the medium-band photometry of 3500 galaxies from the NEWFIRM survey to constrain star formation histories, $0.5 < z < 2.0$. One of our goals here is to determine how competitively PAUS can be used to determine spectral features of galaxies, compared to the use of higher resolution spectra e.g. from zCOSMOS (Lilly et al. 2007), allowing for any modifications to the definitions of the spectral features driven by the narrow-band photometry and taking into account errors in the photometry and in the photometric redshift estimation.

Here, we use the galaxy formation model GALFORM introduced by Gonzalez-Perez et al. (2014), combined with a large-volume, high-resolution N-body simulation to build a mock catalogue for PAUS. Contreras et al. (2013) demonstrated that semi-analytical models of galaxy formation give robust predictions for galaxy clustering and, where differences exist between the models, they can be traced back to choices made in the treatment of galaxy mergers and the spatial distribution of satellite galaxies (see also Pujol et al. 2017). Farrow et al. (2015) used the Gonzalez-Perez et al. (2014) model to interpret GAMA clustering measurements as a function of luminosity, stellar mass, and redshift.

The layout of this paper is as follows. Section 2 introduces the galaxy formation model and the PAUS mock catalogue, Section 3 investigates the use of the PAUS narrow-band filters to measure galaxy spectral features, and Section 4 gives predictions for the narrow-band luminosity functions, other characterizations of the galaxy population in PAUS, and galaxy clustering. We conclude in Section 5.

2 PAUS MOCK LIGHT-CONE

Here, we describe the N-body simulation and galaxy formation model used (Section 2.1), introduce some basic properties of the mock catalogue constructed (Section 2.2), discuss the modelling of emission lines and their impact on narrow-band fluxes (Section 2.3), and set out the treatment of errors in photometry and in photometric redshift errors.

2.1 N-body simulation & galaxy formation model

To model the galaxy population observed with PAUS, we use the GALFORM semi-analytic galaxy formation model presented in Gonzalez-Perez et al. (2014; hereafter GP14). The GALFORM model (Cole et al. 2000) aims to follow the formation and evolution of galaxies in dark matter haloes by solving a set of differential equations that describe the transfer of mass and metals between reservoirs of hot gas, cold gas, and stars (see the recent extensive description of the model by Lacey et al. 2016 and the reviews by Baugh 2006 and Benson 2010). Due to the complexity and uncertainty of galaxy formation physics, many processes are modelled using equations that require parameter values to be specified. These are set by requiring the model to reproduce a selection of observations of the galaxy population, mostly at low redshift. The model calculates the star formation and merger history for each galaxy, including all of the resolved progenitors. With an assumption about the stellar initial mass function and a choice of stellar populations synthesis (SPS) model, GALFORM outputs the flux for each galaxy in the PAUS bands using the composite stellar population obtained from the star formation history (Gonzalez-Perez et al. 2013). This includes a calculation of the attenuation in each band based on the optical depth calculated from the metallicity of the gas and the size of the disc and bulge components of the galaxy (Gonzalez-Perez et al. 2013).

To build a mock catalogue on an observer’s past light-cone with spatial information about the model galaxies, it is necessary to implement the galaxy formation model in an N-body simulation. The dark matter halo merger trees used in the galaxy formation model are also extracted from the N-body simulation (Jiang et al. 2014). The GP14 model is implemented in the Millennium WMAP7 N-body simulation (hereafter MR7, Guo et al. 2013). The MR7 run has a halo mass resolution of $1.86 \times 10^{10} h^{-1} M_{\odot}$ in a cube of side $500 h^{-1}$ Mpc. The use of the MR7 run means that the GP14 model is complete to $i < 23$ for $z > 0.2$. This is sufficient for our analysis. GP14 is an update of the model presented in Lagos et al. (2011) to make it compatible with the WMAP7 cosmology and includes the improved star formation treatment implemented by Lagos et al. (2011).

2.2 Mock catalogue on the observer’s past light-cone

The depth of PAUS means that the properties and clustering of galaxies will evolve appreciably over the redshift range covered. Hence, it is necessary to take this into account when constructing a mock catalogue for PAUS. The starting point is the galaxy population calculated using GALFORM at each of the N-body simulation outputs. Following the light-cone interpolation described in Merson et al. (2013), we construct a mock catalogue of one contiguous 60 deg^2 patch. PAUS will target multiple fields, but this will make little difference to one point statistics and small-scale clustering results presented here.

It is important to demonstrate that the mock catalogue is in broad agreement with the currently available observational data. The number counts of the PAUS mock compare well with large area photometric surveys as shown by Fig. 2, which shows the agreement between the model and the observations from Pan-STARRS (Metcalf, private communication) and the SDSS (York et al. 2000). The systematic differences between the data points are partly due to the slightly different *i*-band filters used in each survey. The offset between the mock catalogue and the data is reasonable when considering the systematic differences between the data. The low-

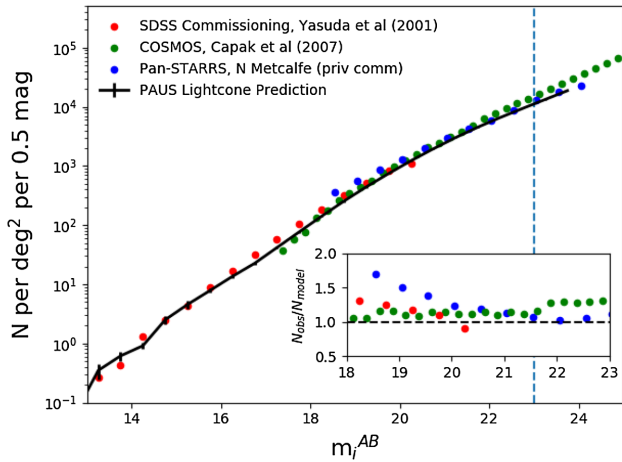


Figure 2. The predicted i -band galaxy number counts in the PAUS mock catalogue (solid line) compared with various observations (coloured symbols; see legend). The vertical bars on the solid line show a jackknife estimate of the sample variance on the number counts. We have omitted the errors on the observational estimates of the counts as they come from very different solid angle surveys. The vertical blue dashed line indicates the PAUS magnitude limit $i = 23$. The inset shows, on a linear scale, the result of dividing the observed number counts by the light-cone predictions.

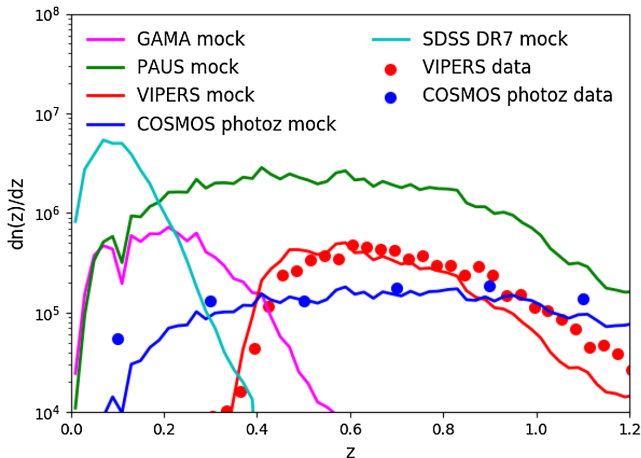


Figure 3. The redshift distributions in various mock catalogues (lines) compared to survey data (circles; see legend). The VIPERS data are taken from de la Torre et al. (2013), and the VIPERS mock catalogue is a 24 deg^2 light-cone to $i < 22.5$ with a 65 per cent sampling rate. The mock VIPERS $n(z)$ is then statistically corrected for the colour cut using the empirical relation found in de la Torre et al. (2013). The COSMOS photo- z data are taken from Ilbert et al. (2009), and the COSMOS photo- z mock is a 2 deg^2 light-cone retaining galaxies with $21.5 < i < 24.5$. The SDSS mock is a $10\,000 \text{ deg}^2$ light-cone with $r < 17.77$, and the GAMA light-cone covers 180 deg^2 to $r < 19.8$. These are plotted without an observational comparison to show the relative survey sizes and depths.

redshift incompleteness due to finite halo mass resolution of the WM7 simulation does not impact this comparison as the total number of faint objects is dominated by galaxies with $z > 0.2$, which are well resolved in the model.

Fig. 3 shows the redshift distributions for the mock light-cones associated with five different galaxy surveys, along with data from VIPERS (de la Torre et al. 2013), and COSMOS photo- z (Ilbert et al. 2009). The choice of the two comparison data sets was made to test the mocks against surveys with flux limits on either side

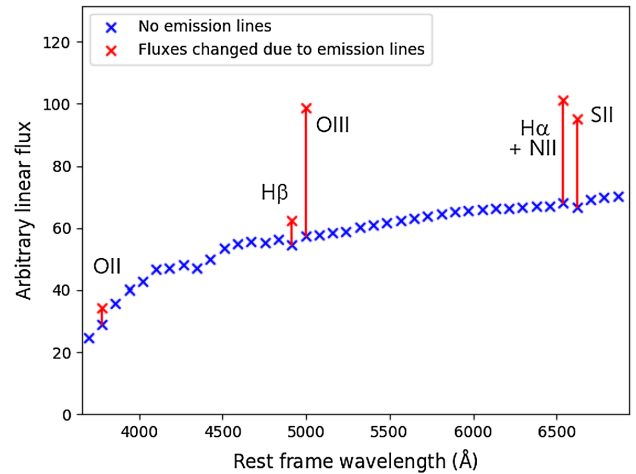


Figure 4. PAUCam filter fluxes for an illustrative star-forming galaxy taken from the PAUS mock. All 40 PAUCam filters are plotted. The blue (red) crosses show filter fluxes without (after including) emission lines.

of the nominal PAUS i -band magnitude limit, VIPERS $i < 22.5$, and COSMOS photo- z with $21.5 < i < 24.5$. The model predictions agree reasonably well with the observations. The disagreement with the lowest redshift COSMOS data point is due to incompleteness in the model; this will be less important for the PAUS mock that is shallower than the COSMOS one. There is some disagreement with the high-redshift tail of the VIPERS $n(z)$. This suggests that the model underpredicts the bright end of the i -band luminosity function at higher redshifts. However, as our analysis is limited to $z < 0.9$, an investigation into the cause and significance of this discrepancy is left to a later date. For $z < 0.9$, the VIPERS mock catalogue agrees well with the observations.

One current limitation of the mock catalogue is that it cannot be used for validation of photometric redshift codes. Tests run using the photo- z code embedded in the PAUS pipeline reveal discreteness in the returned redshifts that are aligned with MR7 snapshots. This issue arises due to the narrow width of the PAUS filters and the associated shift in redshift being smaller than the spacing of the N-body outputs in redshift. This is not an issue for broad-band photometry or when using multiple adjacent filters for measurements as in this analysis. A catalogue constructed using the P-Millennium simulation (Baugh et al., in preparation), will improve both the mass and time resolution of our light-cone mock catalogue.

2.3 Impact of emission lines on narrow-band fluxes

Emission lines are generally thought to make a negligible contribution to the flux measured in broad-band filters, even for high-redshift galaxies (Cowley et al. 2017). However, the narrow width of the PAUCam filters means that it is necessary to revisit the contribution of emission lines for PAUS.

GALFORM makes a calculation of the emission line luminosity of each galaxy using the number of Lyman continuum photons, the metallicity of the star-forming gas, and a model for H II regions from Stasińska (1990). Gonzalez-Perez et al. (2017) give a recent illustration of this functionality presenting predictions for the abundance and clustering of O II emitters.

Fig. 4 shows the contribution emission lines can make to the PAUS narrow-band fluxes for a single model galaxy. This illustrates that emission lines can be beneficial not only for the estimation

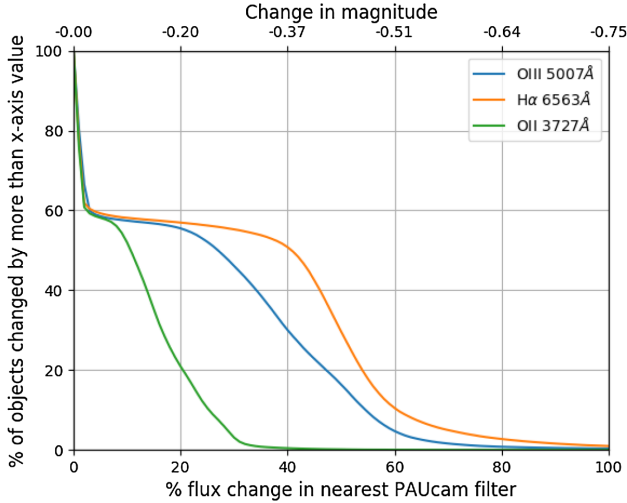


Figure 5. Fraction of model galaxies whose flux in nearest PAUCam filter is affected by the inclusion of a specific emission line (as indicated by the key). Only galaxies with redshift $0.21 < z < 0.3$ and magnitude $i < 23$ are shown to preserve a common sample, where all lines can be sampled by a PAUCam filter. See Section 2.3 for a discussion.

Table 1. Wavelength and redshift ranges over which PAUCam filters (4500–8500 Å) are sensitive to some common spectral features. The table is limited to the main features observable over the redshift range $0.2 < z < 0.9$. See Fig. 7 for the definitions of the PAUS UV and PAUS Blue bands and see Fig. 9 for the definitions of D4000. Note that $M^h \equiv M - 5 \log_{10} h$.

Feature	Wavelength range Å	Redshift range
O II	3727	0.21–1.28
O III	4959/5007	0.0–0.70
H α	6563	0.0–0.29
D4000 _N	3850–3950, 4000–4100	0.17–1.07
D4000 _W	3750–3950, 4050–4250	0.20–1.00
PAUS UV (M_{UV}^h)	3050–3650	0.48–1.39
PAUS Blue (M_B^h)	4050–4450	0.11–0.90

of photometric redshifts but suggests that PAUS could be used to identify and characterize populations of emission line galaxies. This is particularly relevant for the preparations for upcoming large spectroscopic surveys such as DESI (DESI Collaboration 2016) and Euclid (Laureijs et al. 2011) that will build redshift catalogues from emission line galaxies.

Fig. 5 shows the fraction of galaxies whose relevant PAUS filter flux changes by a given percentage due to the contribution of one of the H α , O II, or O III emission lines. For this calculation, we restrict ourselves to a redshift range over which all lines are visible in the PAUCam filter wavelength range (see Table 1). The curves show the change in the flux of the filter with peak transmission closest to the observed emission line. Note that as PAUS filters have an FWHM 130 Å, a full width of ~ 135 Å, and are spaced by 100 Å, in a good fraction of cases a line will also contribute significantly to a second narrow-band flux measurement.

It can be seen from Fig. 5 that for 50 per cent of galaxies in this sample that at least one narrow-band flux measurement changes by 40 per cent or more due to the inclusion of emission lines. That fraction falls to 38 per cent for O III and to 5 per cent for O II due to

the average lower luminosity in these lines compared to that in the H α line.

2.4 Photometry and redshift errors

Photometric redshift errors and photometry errors are added to the mock catalogue in post-processing. Two light-cones are produced, one with perfect photometry and correct redshifts and the other with PAUS-like errors applied. These errors are defined as Gaussian redshift errors of $\sigma_z = 0.35$ per cent and Gaussian flux errors equivalent to a signal-to-noise ratio of 3 at magnitude 23 in the narrow-band filters. These redshift errors are a simple approximation to PAUS photo- z measurements that will be fully explored in Eriksen et al (in preparation). No photometry errors are included in the broad-band magnitudes as the sources of the broad-band photometry will be at least one to two magnitudes deeper than the nominal depth of PAUS of $i = 23$.

Fig. 6 shows the spatial distribution of galaxies in the PAUS mock catalogue and illustrates the impact of different redshift errors on the appearance of the large-scale structure of the universe traced by galaxies. Also shown in Fig. 6 are the model galaxies that satisfy the selection criterion for the GAMA survey, $r < 19.8$ (Driver et al. 2011; plotted at their spectroscopic redshift using blue points). The left-hand panel of Fig. 6 highlights how much richer structures will be in a spectroscopic PAUS compared with GAMA, due to the deeper flux limit. The middle panel of Fig. 6 shows that a significant amount of radial information is retained once the redshifts of the mock galaxies are perturbed by the photometric redshift errors expected for PAUS. At $z \sim 0.3$, the expected photometric redshift errors for PAUS, σ_z of 0.35 per cent, correspond to a comoving distance error of $\sim 13 h^{-1}$ Mpc. Hence, it will be feasible to extract information about group and cluster membership from PAUS (for an example of group finding in a catalogue with less-accurate photometric redshifts than those expected in PAUS (see Jian et al. 2014)). The right-hand panel shows how little radial position information is retained when applying the photometric errors expected for broad-band photometry.

3 PAUS GALAXY PROPERTIES

The PAUS narrow-band filters cover the wavelength range from 4500 to 8500 Å in which certain spectral features can be observed. Over the range in which PAUS will make the greatest contribution to clustering measurements, $0.25 < z < 1.0$, the rest-frame wavelengths from 3000 to 4470 Å are always accessible with PAUS photometry. Table 1 lists the spectral features in the PAUS wavelength range that are investigated here. We assess the direct observation of these features given a galaxy with PAUS-like uncertainties in photometry and redshift. By direct observation, we mean that we calculate the value of a feature by integrating over the appropriate PAUS filter fluxes, assuming that a redshift (of appropriate accuracy) has been measured by the photometric redshift code. An alternative approach would be to extract the spectral information by integrating over the appropriate range of the best-fitting template spectral energy distribution (SED) obtained as part of the photometric redshift estimation. Using the templates in this way could reduce the statistical error, as this approach uses information from all of the filters that are available for a given galaxy. However, this would introduce a systematic error through restricting the results to be derived from combinations of a limited number of templates. It will in fact be best to switch to using templates for measurements

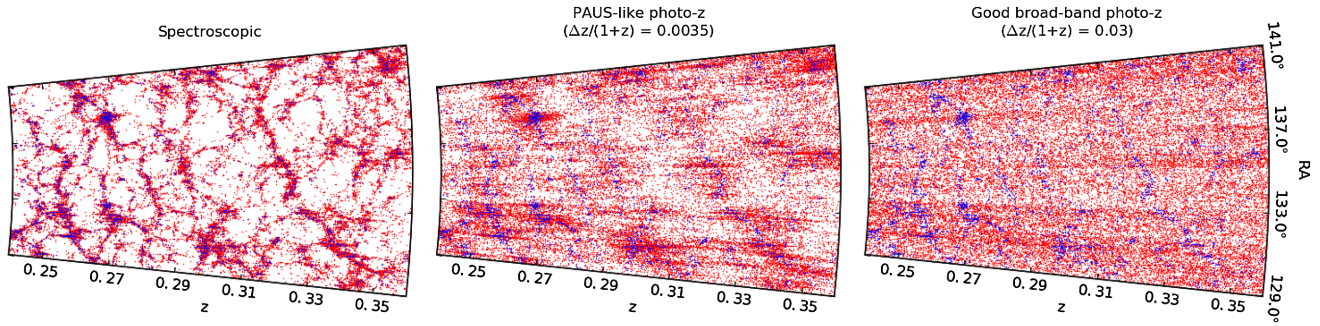


Figure 6. The spatial distribution of galaxies in a 1 deg thick slice from the PAUS mock catalogue. The three panels show the spatial distribution with spectroscopic redshift resolution (left), with PAUS-like redshift resolution (centre) and for typical broad-band redshift resolution (right). The red points are galaxies brighter than the PAUS magnitude limit $i = 23$, while the blue points correspond to GAMA galaxies ($r < 19.8$) with the spectroscopic redshift.

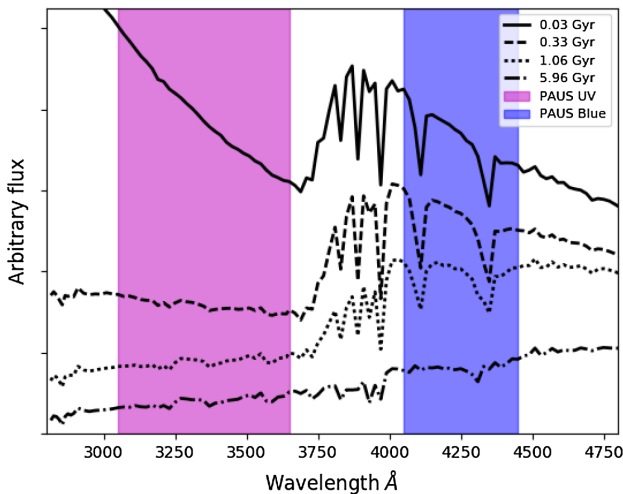


Figure 7. The definition of new rest-frame broad-bands, PAUS UV (magenta), and PAUS Blue (blue). At $z = 0.6$, PAUS UV overlaps with 9.6 PAUCam filters and PAUS Blue overlaps with 6.4 PAUCam filters. The curves shown are some of the SEDs for single age stellar populations that are used in the construction of the mock catalogue. In all cases, these are for one quarter solar metallicity, with ages given in the key.

whose statistical errors exceed a certain threshold. The exact threshold is unknown as it depends on the unquantifiable systematic of template incompleteness, but this analysis can be used to define the point at which direct measurements become unfit for purpose, i.e. when must we switch to using templates.

3.1 Rest-frame defined broad-bands

We define rest-frame broad-bands to best utilize the narrow-band information from PAUS. These quantities are calculated by integrating the interpolated low-resolution spectrum provided by the narrow bands. This type of direct rest-frame measurement is possible because each of the PAUCam filters is flux calibrated, something which is often not the case with higher resolution spectra.

As can be seen from Fig. 7 and Table 1, the PAUS UV band has been chosen to be bluewards of the 4000 Å break, and hence is sensitive to very young stars in the composite stellar population of a galaxy. Conversely, the PAUS Blue band is chosen to be redwards of the break, and thereby probes somewhat older stellar content. PAUS UV is chosen to be wider than PAUS Blue to increase its signal-to-noise ratio. This is important particularly for the UV band

due to the typical shape of an i -band selected galaxy SED meaning that, on average, the UV is fainter than the Blue. PAUS Blue can only be directly measured up to $z = 0.9$.

There are several benefits to using these new rest-frame broad-bands over and above single narrow bands or traditional broad-bands:

- (i) These bands cover multiple narrow-band filters, increasing the signal-to-noise ratio of an individual measurement, compared to using a single narrow band.
- (ii) They are near direct measurements of galaxy rest-frame SEDs and so do not require average k -corrections that broad-band colour selections often require.
- (iii) They can be chosen to sample desirable sections of a galaxy SED precisely.
- (iv) Similar analyses can be performed on other photometrically calibrated spectra.
- (v) The filter wavelengths are fixed in the observer frame but sample a wavelength range in the rest frame that shrinks as $1/(1+z)$ with increasing redshift. This means that the rest-frame magnitudes we have defined are measured using filters that become more closely spaced as the redshift of the source increases. Hence, the rest-frame magnitudes are better sampled with increasing redshift, which partly offsets the typical decrease in the signal-to-noise as sources get fainter.

Fig. 8 shows how PAUS redshift and photometry errors propagate into errors in the PAU UV and PAU Blue magnitudes for a sample of mock galaxies with redshifts in the range $0.5 < z < 0.63$ and $i < 23$. For 80 per cent of model galaxies at $i = 23$, PAUS Blue can be measured to within ± 0.2 magnitude and PAUS UV to within ± 0.25 magnitude. There is also no bias in the measurement at all values of i -band magnitude. Other redshift selections give similar errors and also show no bias.

3.2 The 4000 Å break

The 4000 Å break is driven by a combination of Ca II absorption lines and CN bands in the spectra of old stars. The quantity D4000 is the ratio of average flux in one spectral region at wavelengths just above 4000 Å and that in a region just below in wavelength. The literature defines this quantity in two ways, D4000 narrow defined in Balogh et al. (1999) and D4000 wide defined in Bruzual (1983). The two flux bands used are different in each case and are visualized in Fig. 9. We first investigate if PAUCam has high enough resolution in a high-signal-to-noise scenario to measure D4000 wide and narrow

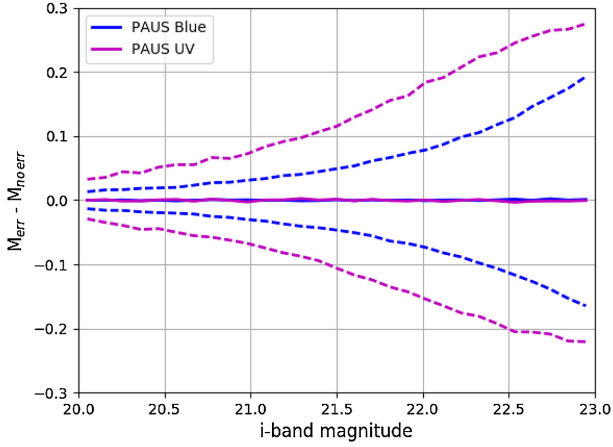


Figure 8. Statistical uncertainty in the PAUS UV and PAUS Blue magnitudes as a function of i -band magnitude for mock galaxies with $0.5 < z < 0.63$. The uncertainty includes redshift and photometry errors as described in Section 2.4. The solid lines show the median error, and the dashed lines show the 10–90 percentile range.

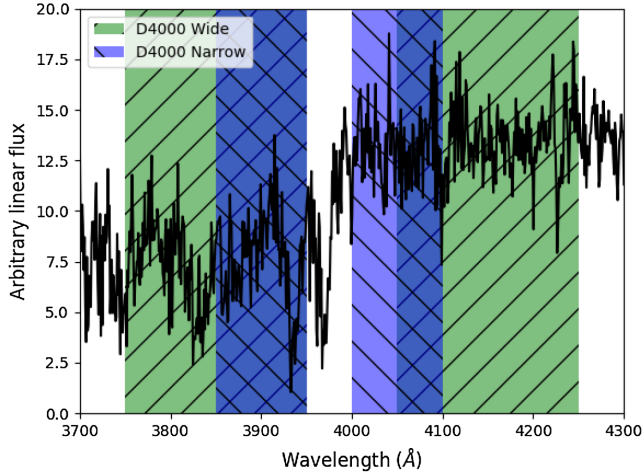


Figure 9. Definitions of D4000 wide and D4000 narrow overlaid on a randomly selected, de-redshifted, SDSS DR10 galaxy. The green shaded region represents the wide definition (3750–3950 and 4050–4250 Å) from Bruzual (1983), and the blue the narrow (3850–3950 and 4000–4100 Å) from Balogh et al. (1999).

and then separately investigate D4000 measurements of PAUS mock galaxies.

3.2.1 Measuring the 4000 Å break strength with PAUCam spectral resolution

In order to test the measurement of the D4000 feature, we look at a sample of 4500 SDSS DR12 galaxy spectra, selected around $z = 0.1$ (Smee et al. 2013; Alam et al. 2015). We consider SDSS spectra for this test as the SPS used in GALFORM is limited to 20 Å resolution. The SDSS galaxies were each randomly uniformly placed at a redshift in the range $0.2 < z < 0.9$, so that the different ways in which the PAUS filter can trace the feature are taken into account. The fluxes in the 40 PAUCam narrow bands were calculated for each galaxy. D4000 was then calculated using both the full-resolution SDSS spectra, and then again by integrating a linear interpolation of the PAUS filter measurements. Both definitions of D4000 from

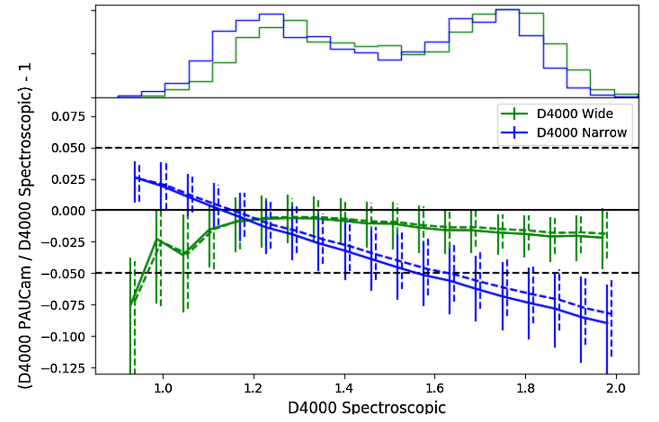


Figure 10. Relative accuracy with which D4000 can be recovered using PAUCam, as a function of the strength of D4000, measured using 4500 SDSS spectra observed at $z \sim 0.1$ and redshifted over the interval $0.2 < z < 0.9$. D4000_s is measured using the full spectra information, while D4000_p uses the PAUS filters. The green line shows the result for D4000 wide and the blue for D4000 narrow. The solid lines and error bars (which indicate the 10–90 percentile range) include a PAUS-like photo- z error, while the dotted lines and error bars do not. The dotted lines are displaced in the x direction by 0.01 to make the error bars visible. The top panel shows the distributions of D4000 values for the sample.

the literature were calculated and results are presented with and without PAUS-like redshift errors, as defined in Section 2.4. We do not include photometry errors, as first we want to check if PAUCam has sufficient resolution to measure D4000 in a high- signal-to-noise scenario.

Fig. 10 shows how well interpolating between the PAUCam filters recovers the spectroscopic result for both the wide and narrow D4000 definitions from the literature. Both definitions of D4000 are biased due to the effective smoothing of a sharp spectral feature due to the finite-width wavelength intervals used to calculate D4000. D4000_n is affected by this bias more than D4000_w. The D4000_n bias also scales as a function of the spectroscopic value for D4000, whereas the bias of D4000_w is nearly constant with respect to this ideal. The D4000_w measurement is biased by ~ 2 per cent. This bias is not corrected for in later analysis as we will see in Section 3.2.2 that it is small compared to the random errors on PAUS mock galaxies. Once photometric redshift errors are included, the error bars on both measurements increase only slightly. The error bars on D4000_w are also smaller than those of D4000_n, $\sim \pm 2$ per cent and $\sim \pm 4$ per cent, respectively. The superior recoverability of D4000_w suggests this 4000 Å break definition should be used for PAUS measurements. The superior bias and noise performance of D4000_w is to be expected as it overlaps with more PAUCam filters than D4000_n does at a given redshift.

The redshift dependence of the D4000_w measurement bias was investigated, as at each redshift the filters will trace the break in a different manner. The extreme scenarios are that the D4000 break lies mid-way across a filter or exactly in between two filters. It was found that the bias of D4000_w varies by less than 1 per cent as a function of redshift. It is therefore not necessary to model this redshift dependence.

3.2.2 4000 Å break strength in PAUS

To investigate the ability of using the PAUS photometry to measure D4000_w, this quantity is measured in both the mock catalogue with

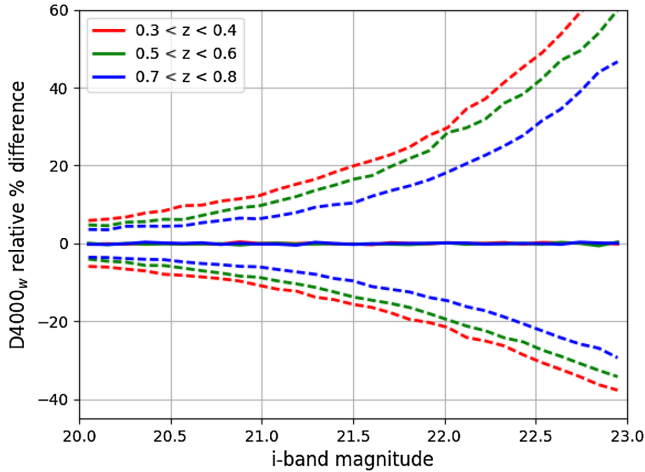


Figure 11. Relative percentage difference in $D4000_w$ as a function of i -band magnitude for different redshift slices. The relative percentage difference is defined as $100 \times (D4000_{\text{err}} - D4000_{\text{true}})/D4000_{\text{true}}$, where the subscript $\text{err}(\text{true})$ refers to measurements made in the catalogue with (without) PAUS simulated redshift and photometric errors.

no errors and in the one with redshift and photometric errors introduced in Section 2.4. Fig. 11 shows the relative error in $D4000_w$ for redshift slices as a function of i -band magnitude. 80 per cent of galaxies at $i = 23$ lie within 50 per cent of the true value of $D4000_w$. Photometric uncertainty is therefore the dominant source of error for PAUS galaxies. Looking at the population histogram in Fig. 10, it can be seen that the majority of galaxies have values of $D4000_w$ between 1.0 and 2.0, with a bimodal distribution peaking at 1.2 and 1.75. An error of 50 per cent is therefore very large compared to the range of $D4000_w$. Galaxies with $i = 21.5$ and $z = 0.55$, however, are expected to have just a 15 per cent error in $D4000_w$, showing that direct $D4000_w$ measurements for a bright subset of PAUS objects are feasible. $D4000_w$ errors are smaller for higher redshift galaxies at a fixed i -band magnitude as the rest-frame defined $D4000_w$ bands overlap with more PAUCam filters in this case than at lower redshifts. Individual studies will need to define a tolerable error for this quantity. Bimodal population cuts, for example, will be able to use a large subset of data and retain completeness and purity, whereas studies on the ages of individual galaxies may need to use a significantly restricted subset of the catalogue. One could also stack populations of galaxies and make a measurement on mean spectra to reduce statistical error.

4 RESULTS

In this section, we review various properties of the galaxy population that we expect PAUS will be able to measure based on the predictions made using our mock catalogues.

4.1 Narrow-band luminosity functions

The parameters in the GALFORM model are calibrated to match low-redshift observations, which are mainly one-point statistics such as the luminosity function. One of the applications of PAUS is to provide improved constraints on the model parameters by providing measurements of the narrow-band luminosity function over a significant baseline in redshift.

We have seen that individual PAUCam narrow-band magnitudes can be significantly affected by the emission line flux from a galaxy,

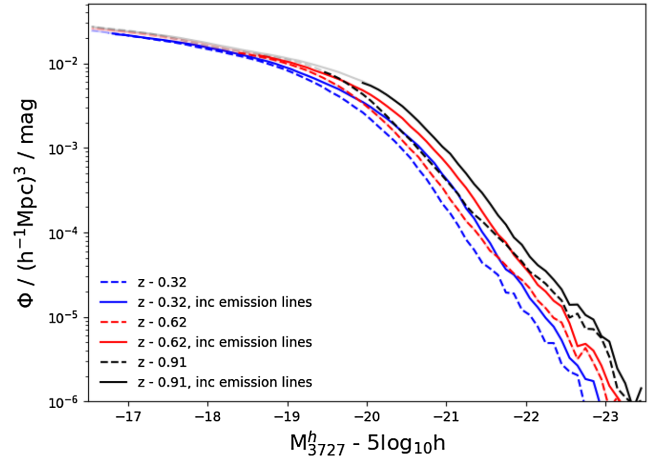


Figure 12. Luminosity functions at several snapshot redshifts (as labelled) of a PAUS filter at rest-frame wavelength of 3727 ± 62.5 Å. A different PAUS filter is used at each redshift, chosen to overlap with the O II emission line. The solid lines show the prediction including the emission line flux and the dashed lines do not. The plotted curves become fainter when they fall below 95 per cent completeness at $i < 23$.

so here we investigate the sensitivity of the narrow-band luminosity functions to the inclusion of emission lines in the GP14 model (see Gonzalez-Perez et al. 2017 for a further discussion of model predictions for O II emitters). Fig. 12 shows how a narrow-band luminosity function of PAUCam-like filter chosen to overlap in the rest frame with the O II emission line changes when the flux from the line is included. Measurements are made in the simulation snapshots. Inference of this quantity from observer frame measurements would require accurate k -corrections and is beyond the scope of this paper. It can be seen that neither redshift evolution nor inclusion of emission line flux change the faint-end slope of the luminosity function in the GP14 model. The value of M^* , however, increases with both redshift, as a result of the increasing star formation, and also with the inclusion of O II line flux. The contribution of the stellar continuum to the flux in this band can be estimated by averaging the flux in bands placed at either side of the band that contains the O II emission, providing a constraint on the amount of emission line flux and its evolution with redshift.

4.2 Characterization of the galaxy population

One desirable objective for studying the evolution of the galaxy population is the ability to separate galaxies by colour in a consistent way across the redshift range sampled by PAUS. This objective can be achieved using a cut in $D4000$ at $z < 0.5$ and a cut in $M_{\text{UV}}^h - M_{\text{B}}^h$ above redshift 0.5. We could define a band further into the red to make a colour cut at lower redshifts, as M_{UV}^h cannot be defined below $z \sim 0.5$, see Table 1, but a cut in a different section of a galaxy SED may non-trivially select galaxies differently than the $M_{\text{UV}}^h - M_{\text{B}}^h$ cut. In particular, the use of a redder colour selection might mix galaxies with different recent star formation histories, making clustering comparisons across redshift ranges less informative. The use of $D4000_w$ means that we are making a colour cut centred on the same portion of the SED as a cut in the colour $M_{\text{UV}}^h - M_{\text{B}}^h$.

Fig. 13 shows the distributions of $D4000_w$ and $M_{\text{UV}}^h - M_{\text{B}}^h$ for a redshift range in which both can be measured. Both quantities show a bimodal distribution, which we can loosely refer to as ‘red’ and

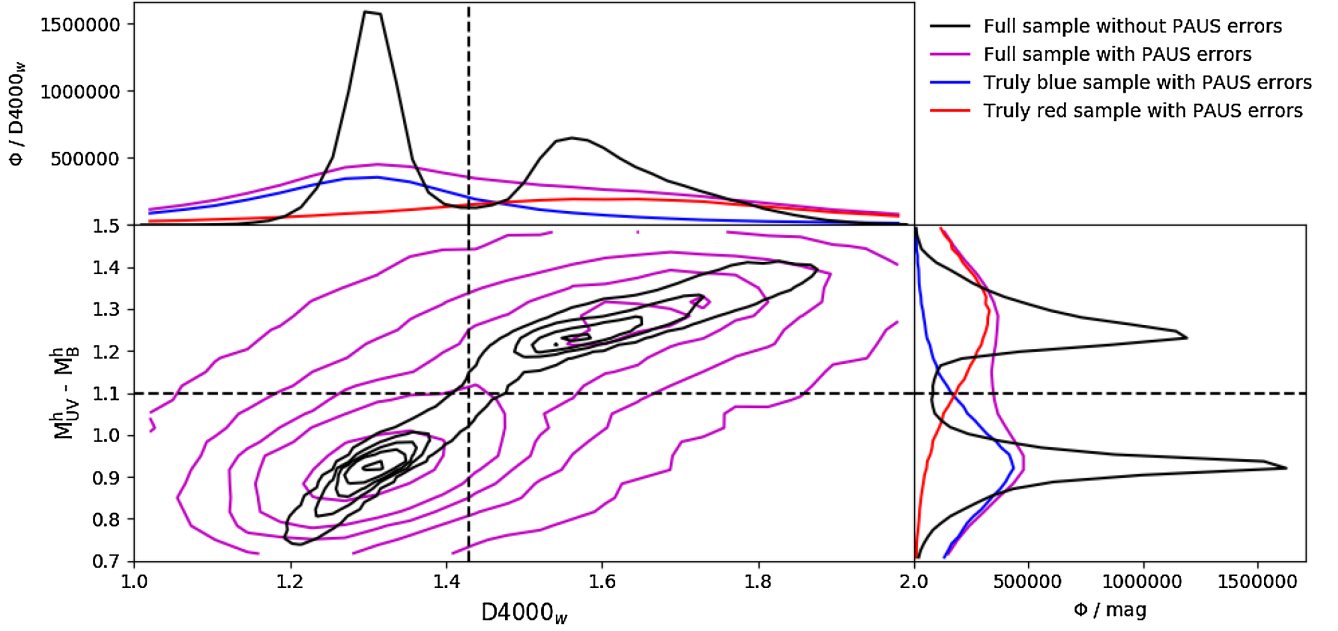


Figure 13. Distribution of galaxies with $i < 23$ and $0.5 < z < 0.63$ in the $D4000_w$ and $M_{UV}^h - M_B^h$ colour plane, with and without simulated PAUS errors. The contours contain 10, 30, 50, 70, and 90 per cent of the sample. The solid black lines show the distributions for the full sample without errors, and the magenta ones show the full sample with errors. The red (blue) curves show the distribution of galaxies that are intrinsically red (blue) in each measure when errors are included.

‘blue’ populations. A cut is made at $D4000_w = 1.42$ and $M_{UV}^h - M_B^h = 1.1$. Before photometric errors are added, disagreements in red–blue classification when using the two measures are at the subpercent level. The cut in $M_{UV}^h - M_B^h$ is appropriate to split the bimodal population at higher redshifts, as is the cut in $D4000_w$ for lower redshifts. Comparisons carried out using the model rest-frame bands show that these colour cuts are similar to a traditional broad-band rest-frame cut in $u-g$. When including photometric errors, mixing between the red and blue populations is more severe when using $D4000_w$ than with the rest-frame magnitudes due to the larger fractional error in $D4000_w$ at a fixed i -band magnitude (see Sections 3.1 and 3.2.2). Errors on the $M_{UV}^h - M_B^h$ colour are driven largely by errors in the UV magnitude.

4.3 Galaxy clustering

We select volume-limited galaxy samples for clustering measurements based on redshift, PAUS Blue luminosity and rest-frame colour (as defined in Section 4.2). We choose not to split samples based on inferred quantities such as star formation rate or stellar mass as the inference of these properties from narrow-band photometry is left to future work. Inferring these properties has also been shown to introduce biases based on the assumptions made in these inferences (Mitchell et al. 2013). In the mock including simulated PAUS errors the cuts are made after all sources of error are included. See Appendix A for clustering definitions, details of the calculations and open source code links, and Appendix B for more information on sample selection. All errors in this section are calculated using a jackknife over 12 regions in the simulated survey, see e.g. Norberg et al. (2009).

We estimate the galaxy bias from the ratio of the projected galaxy clustering to the projected clustering of the MR7 dark matter at the median redshift of the sample in question. The values of the correlation function for the MR7 snapshots were taken from McCullagh

et al. (2016). This quantity allows us to separate the evolution of the dark matter overtime from the evolution of the galaxy population. On large scales, this quantity is equal to the linear bias. More specifically, we define projected galaxy bias as

$$b(r_p, z) = \sqrt{\frac{w_p(r_p, z)}{w_p(r_p, z)_{DM}}}, \quad (1)$$

where $w_p(r_p, z)$ is the projected correlation function defined in equation (A4).

4.3.1 Impact of photometric uncertainty

Fig. 14 shows the bias measured for one mock PAUS sample ($-19.5 < M_B^h < -19.0$) in the redshift range $0.5 < z < 0.63$, both with and without PAUS magnitude and photometric redshift errors. The value of π_{\max} used was $100 h^{-1}$ Mpc. Fig. A1 in the Appendix shows the recovery of the projected correlation as a function of different photometric redshift errors. A value of π_{\max} of $50 h^{-1}$ Mpc would have been sufficient for the photometric redshift errors assumed in this work, and would have slightly reduced the statistical noise, but the real survey will have a distribution of photometric redshift errors, so the conservative value of $100 h^{-1}$ Mpc was chosen.

For the sample selected only on redshift and M_B^h , the black lines in Fig. 14, the projected clustering signal is recovered without systematic error when including PAUS-like errors. The jackknife statistical errors only slightly increase when compared with the ideal case. This demonstrates that the PAUS photo- z measurements are sufficient to calculate the projected galaxy clustering without systematic error. Table B1 in the appendix shows that the sample with PAUS-like errors is over 90 per cent pure and complete. For the same sample with photo- z errors only and no photometry error, these numbers both rise above 96 per cent, showing that mixing between samples due to photometric redshift errors is minimal.

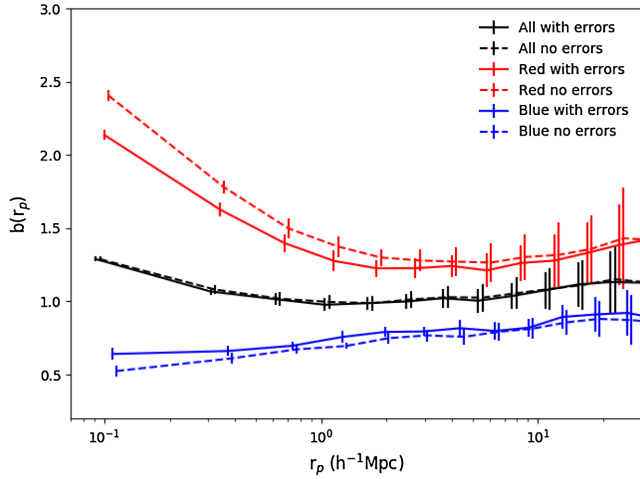


Figure 14. Projected galaxy bias (equation 1) for a typical PAUS sample ($0.5 < z < 0.63$, $-19.5 < M_B^h < -19.0$). The full galaxy sample is shown in black, and the results on splitting the sample into red and blue populations are shown in these colours. The solid lines show the results using the light-cone with redshift and photometry errors taken into account and the dashed lines the results without including these uncertainties. Errors calculated using jackknife resampling.

Once a colour cut is applied to the full magnitude-limited galaxy sample, a significant difference can be seen in the projected bias measurements for the red and blue populations. Errors in the photometry introduce mixing between the red and blue populations that leads to a small reduction in the difference between the one-halo scale projected bias of red and blue galaxies. Nevertheless, the difference between the clustering measurements for these populations remains significant. Systematics on two-halo scales are within the statistical uncertainties. This confirms that the most significant source of systematic error in this analysis will be on one-halo scales and come from the misclassification of galaxies into red or blue subsamples using these direct rest-frame measurements. This systematic error shows up here as there is a large contrast between the one-halo clustering of red and blue samples, and PAUS will have small statistical errors on those scales. Again, statistical colour errors could be reduced using the best-fitting photo- z SED inferred colours for fainter samples, but this is not tested here. This highlights the importance of understanding sample selection and the role of mock catalogues in interpreting clustering results.

4.3.2 The redshift evolution of clustering

Fig. 15 shows the predicted redshift evolution of projected galaxy bias measured for samples of red and blue galaxies with $-19.5 < M_B^h < -19.0$. Our estimate of the bias naturally takes into account the evolution of the clustering of the dark matter over this redshift interval. For all redshift bins, red galaxies show stronger clustering than blue galaxies. This difference becomes larger for pair separations below $\sim 1 h^{-1}$ Mpc corresponding to pairs within common dark matter haloes. The bias also increases with redshift for both red and blue samples. This trend is also seen in all the other luminosity bins we have explored. This result, the decline in the bias as the universe ages, is due to faster growth of the dark matter correlation function compared with that of the galaxy correlation function over the same period, see e.g. Baugh et al. (1999). Again the systematic errors on two-halo scales are within statistical uncertainties. Quali-

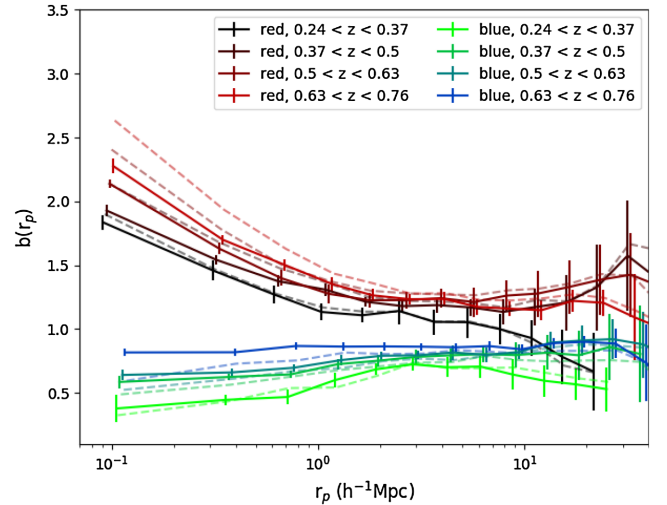


Figure 15. Projected galaxy bias (equation 1) inferred from the projected correlation function measured for samples with $-19.5 < M_B^h < -19.0$, split by colour and redshift. The solid lines show the results using the light-cone including redshift and photometry errors, and the dashed lines show the results without these uncertainties. Errors, from jackknife resampling, only shown for PAUS-like sample.

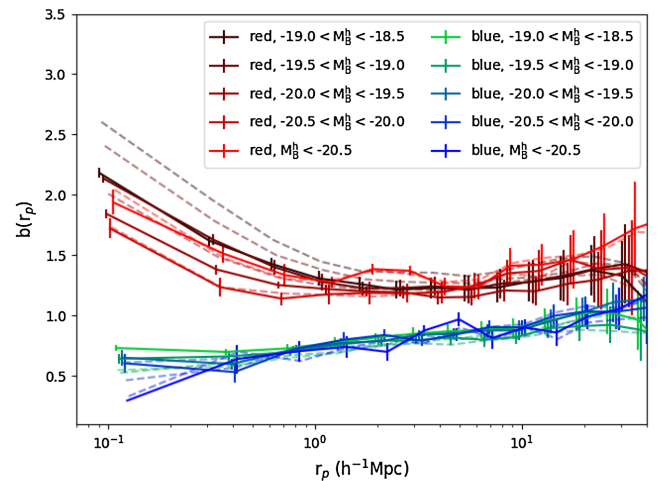


Figure 16. Projected galaxy bias (equation 1) inferred from the projected clustering measured for samples $0.5 < z < 0.63$, split by colour and M_B^h . Line types as in Fig. 15.

tative trends seen on one-halo scales are preserved once errors are included, but the contrast between red and blue one-halo clustering is reduced due to colour mixing.

4.3.3 The luminosity dependence of galaxy clustering

Fig. 16 shows the model prediction for the luminosity dependence of galaxy clustering. The split between the red and blue galaxies is once again very evident. As commented above, the red samples have stronger clustering than their blue counterparts. There is little luminosity dependence of the clustering measure for the blue samples (see also Kim et al. 2009 for a discussion of the luminosity dependence of clustering in an earlier version of the GALFORM model used here). On the other hand, the clustering of the red samples shows a moderate dependence on luminosity that weakens on large

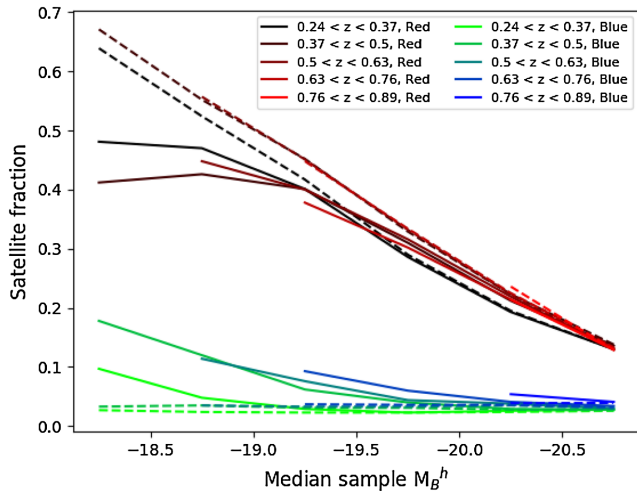


Figure 17. Satellite fraction as function of M_B^h for galaxy samples split by colour and redshift. Line types as in Fig. 15.

scales and does not preserve the same ordering with luminosity that is displayed on small scales. Once again two-halo scale results are recovered within statistical errors.

One reason for the inverted trend of clustering decreasing with luminosity seen on small scales is due to the dominance of satellite galaxies in the lower luminosity red samples. This can be seen in Fig. 17, which shows the satellite fractions of the clustering samples (Number of galaxies with satellite label in a sample divided by the total number of galaxies in the sample). Note that measuring this with the data would require significant modelling work. This figure also illustrates the impact of colour mixing on the satellite fraction of the samples. The lower luminosity bins at the lowest redshifts are significantly affected by mixing between centrals and satellites. These lower luminosity and redshift samples have the largest difference in satellite fraction between the red and blue populations and are the most likely to be misclassified in colour. This mixing error will either need to be modelled using mocks or we will have to rely instead on inferred colours extracted from an SED template, allowing for template incompleteness as a systematic error.

5 CONCLUSIONS

We have introduced a mock catalogue built from a semi-analytical model of galaxy formation implemented in an N-body simulation for use in conjunction with the PAUS. PAUS is a novel narrow-band imaging survey that is underway on the William Herschel Telescope. The width of the PAUS filters means that photometric redshifts of unprecedented accuracy will become available for a homogeneously selected sample of galaxies down to $i = 23$. The PAUS mock is built using the GP14 GALFORM model (Gonzalez-Perez et al. 2014), which is run on the MR7 N-body simulation (Guo et al. 2013). The galaxy snapshots produced at the output times of the MR7 run are then used to construct a mock catalogue on an observer’s past light-cone, which predicts the evolution of the clustering of galaxies and their properties (Merson et al. 2013). The mock catalogue is available on request at CosmoHub¹ (Carretero et al. 2017).

The resulting mock catalogue agrees with observed galaxy number counts to within the scatter between different surveys. Over the redshift range in which PAUS is expected to make the largest impact, $0.2 < z < 0.9$, the mock is in good agreement with the redshift distributions from COSMOS photo- z and VIPERS. There is some tension at $z > 1$ where the mock under predicts the VIPERS $n(z)$, but this redshift range is less relevant for PAUS, and the observational errors are large at these redshifts (de la Torre et al. 2013).

We include galaxy emission lines in the predicted PAUS measurements and show that this has a significant effect on PAUS narrow-band fluxes. We show how the rest-frame narrow-band luminosity function changes when emission lines are included by choosing a rest-frame narrow band that overlaps with the O II emission line. The GP14 GALFORM model predicts no change in the faint-end slope of the narrow-band luminosity function with or without emission line flux included and as a function of redshift. It does, however, predict an increase in M^* with both redshift and on the inclusion of emission lines.

We define rest-frame broad-bands calculated directly from narrow-band fluxes and predict that a PAUS Blue (PAUS UV) flux can be directly measured with an error of ± 0.15 (± 0.25 magnitude) down to $i = 22.5$. These provide rest-frame measurements without needing to make any of the assumptions that come with average k -corrections used with broad-band measurements. These rest-frame measurements are only possible because the PAUS narrow-band measurements are flux calibrated. We show that the PAUCam filter set has sufficient resolution to measure the strength of the 4000 Å break, D4000. We predict that D4000_w can be directly measured in PAUS to better than ± 10 per cent precision for galaxies with $i < 21.5$. Providing errors on these quantities as a function of i -band magnitude will allow the PAUS data analysis pipeline to decide when to switch from directly measuring a quantity using the observed PAUCam filters to integrating over the best-fitting SED assigned by a photometric redshift code. The latter incorporates statistical information from all filters but restricts results to a linear combination of SED templates, and is not explored here.

We explore galaxy clustering measurements over a redshift range of 0.2–0.9 for multiple luminosities and colours using the rest-frame colours, D4000_w and redshift. PAUS will provide a unique sample spanning this redshift range over a larger area than previously possible, with nearly 100 per cent completeness. No close galaxy pairs are missed as is often the case in spectroscopic surveys.

We show that systematic errors in projected clustering recovery due to PAUS photometric redshift errors are significantly smaller than statistical errors. All two-halo scale projected clustering results are recovered within statistical errors once PAUS redshift and photometry errors are included. One-halo scale clustering shows the same qualitative trends as measurements made in the ideal case but there is a loss of contrast between the one-halo scale clustering of red and blue galaxies caused by colour misclassification. This demonstrates the importance of a mock catalogue to interpret galaxy clustering results, particularly in the case of PAUS results on small scales, where statistical errors are small and any systematics are likely to be the dominant source of error.

We provide testable predictions for the mock catalogue that the measured galaxy clustering will evolve more slowly with redshift than the redshift evolution in the dark matter, especially for the one-halo term. The mock also predicts that red galaxies will cluster more strongly than blue galaxies. We also predict that fainter galaxies will cluster more strongly than brighter galaxies on small scales due to

¹<https://cosmohub.pic.es/home>

their larger satellite fraction, and that this trend will be particularly strong for red galaxies.

This work provides a tantalizing illustration of the science that will be possible with PAUS, particularly with a view to constraining the galaxy – dark matter halo connection.

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Study, the Japan Participation Group, Johns Hopkins University, the Joint Institute for Nuclear Astrophysics, the Kavli Institute for Particle Astrophysics and Cosmology, the Korean Scientist Group, the Chinese Academy of Sciences (LAMOST), Los Alamos National Laboratory, the Max-Planck-Institute for Astronomy (MPIA), the Max-Planck-Institute for Astrophysics (MPA), New Mexico State University, Ohio State University, University of Pittsburgh, University of Portsmouth, Princeton University, the United States Naval Observatory, and the University of Washington.

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APPENDIX A: GALAXY CLUSTERING STATISTICS AND CODE

We calculate galaxy clustering using the appropriately normalized Landay–Szalay estimator (Landy & Szalay 1993):

$$\xi(r_p, \pi) = \frac{DD(r_p, \pi) - 2DR(r_p, \pi) + RR(r_p, \pi)}{RR(r_p, \pi)}, \quad (A1)$$

where DD, DR, and RR are normalized Data–Data, Data–Random, and Random–Random pair counts, respectively. The number of randoms set is always 10 times the number of galaxies in a sample, and they are uniformly distributed in the comoving volumes of the samples. r_p and π are, respectively, the galaxy pair separations transverse and parallel to the line of sight. These separations are defined in terms of the pair of galaxy vectors \underline{x}_1 and \underline{x}_2 :

$$\pi = \left| \frac{(\underline{x}_1 - \underline{x}_2) \cdot (\underline{x}_1 + \underline{x}_2)}{|\underline{x}_1 + \underline{x}_2|} \right|, \quad (A2)$$

$$r_p = \sqrt{(\underline{x}_1 - \underline{x}_2)^2 - \pi^2}. \quad (A3)$$

In this analysis, we consider only projected galaxy clustering to minimize the impact of the PAUS redshift error. The projected correlation function is given by

$$w_p(r_p) = 2 \int_0^{\pi_{\max}} \xi(r_p, \pi) d\pi, \quad (A4)$$

where the value of π_{\max} is a parameter to be set.

Fig. A1 shows the systematic loss of signal in the projected galaxy clustering for samples with different values of photometric redshift errors relevant to PAUS for two different values of π_{\max} . The sample used was $(-19.5 < M_B^h < -19.0)$ in the redshift range $0.5 < z < 0.63$. The real PAUS data will have a distribution of photometric redshift errors rather than the single Gaussian error

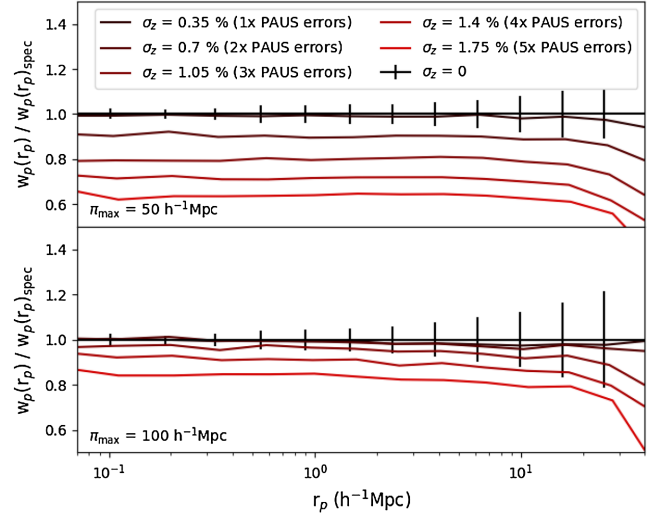


Figure A1. The recovery of the projected galaxy clustering for samples of different Gaussian photometric redshift errors and different values of π_{\max} . Each curve is normalized by the spectroscopic result integrated to the same π_{\max} . The error bars represent the jackknife errors on the spectroscopic result.

assumed here, so this plot can inform us on the systematic errors we may introduce for different error distributions. The larger value of π_{\max} recovers more of the signal but at the cost of increasing the statistical noise. The difference in spectroscopic result between $\pi_{\max} = 50$ and $100 h^{-1} \text{Mpc}$ is less than 2 per cent. A value of π_{\max} of $100 h^{-1} \text{Mpc}$ would allow us to use galaxies in the sample with three times the nominal PAUS redshift error and recover the projected clustering within the statistical errors. See Arnalte-Mur et al. 2009 and Arnalte-Mur et al. 2014 for further discussion on projected-correlation recovery in photometric redshift surveys.

All clustering results are calculated using a two-point clustering code that is publicly available on github.² This is an OpenMP accelerated code that has the ability to calculate monopole and 2D decompositions of the correlation function with flexible linear or logarithmic binning, multiple input/output types, and on-the-fly jackknife errors at the expense of very little extra computing time.

The galaxy pairs were binned logarithmically in both r_p and π , which can help reduce the increase in statistical error for large values of π_{\max} .

APPENDIX B: CLUSTERING SAMPLES

Fig. B1 shows the volume-limited cuts used to create galaxy clustering samples. The faint limit in M_B^h at each redshift was chosen such that the faintest samples were over 99 per cent complete in a catalogue $i < 23$ without errors. The scatter in the colour term between the observed i band and M_B^h is responsible for any small amount of incompleteness. The high completeness of the samples can be seen from Fig. B1 by noting that the bottom right corners of the faintest boxes do not overlap with galaxies with mean i -band magnitude of 23. These samples are therefore the samples we would

²<https://github.com/lstothert/two-pcf>

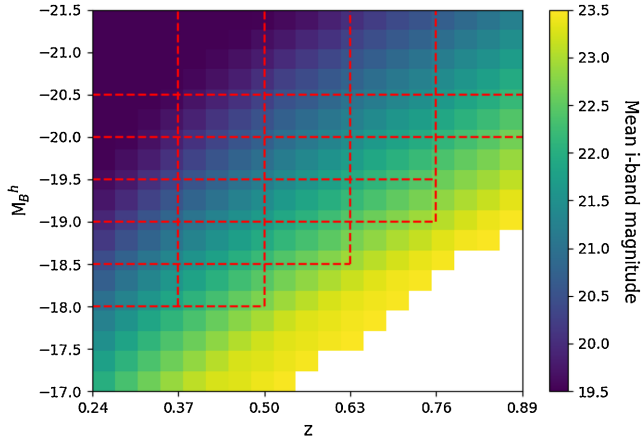


Figure B1. Rest-frame M_B^h versus redshift, colour coded by mean i -band magnitude for a PAUS mock built to $i < 25$ without including redshift or photometry errors. The light-cone was built deeper than nominal PAUS depth so as to be certain about the completeness values of the samples. The plot stops at $i < 23.5$, so the colour gradient through the boxes is more obvious to the reader. The boxes show the sample limits used in the galaxy clustering analysis, chosen to be 99 percent complete to $i < 23$ in this light-cone. Note the boxes do not touch the $i = 23$ coloured squares.

choose if we had perfect photometry, and we then deduce the recoverability of the results when realistic errors are included. The cuts at lower redshift must have a more conservative limit in M_B^h than at higher redshift as the scatter between PAUS Blue and the apparent i -band magnitude is larger at lower redshift. This is because at the

lowest redshift the wavelength difference between the two bands is maximized in the PAUS redshift range so the colour term, and the corresponding colour scatter, is the largest.

All samples selected along with their completeness and purity once errors are included are listed in Tables B1 and B2. The definitions of the completeness and purity in those tables can be written as follows. Define N_{ij} as the number of galaxies that lie in sample i in the catalogue without errors and in sample j in the catalogue including errors. Define N_{i*} as the number of galaxies in sample i in the catalogue without errors. Define N_{*j} as the number of galaxies in sample j in the catalogue with errors. The completeness of sample i can now be defined as N_{ii} / N_{i*} and the purity as N_{ii} / N_{*i} . Satellite fraction and median halo mass are galaxy-weighted quantities. A halo with many satellites may therefore make multiple contributions to the number of satellites and halo masses in a sample. The samples here were split in uniform redshift steps but future work may choose to make the lower redshift bins larger than the higher redshift bins to match the sizes of the volumes probed.

There is high completeness and purity amongst samples split only by redshift and M_B^h seen in Table B1, which drops when samples are further split by colour in Table B2. This shows that the driving source of sample mixing in this work is the colour split. In a fixed luminosity bin, the completeness and purity falls with redshift as the photometry errors are larger for apparently fainter samples. This also holds once galaxies are split by colour.

The number density of the brightest galaxies increases with increasing redshift as the star formation rate of the universe, and therefore the amplitude of the M_B^h luminosity function, increases

Table B1. Table of galaxy clustering samples used in this analysis. Completeness, purity, and satellite fraction are defined in the text. \bar{n} is the number density of the sample.

z-min	z-max	M_B^h bright	M_B^h faint	Comp (%)	Purity (%)	\bar{n}	Sat frac	Median M_{halo}
Volume	$(10^6 h^{-3} \text{ Mpc}^{-3})$					$(10^{-3} h^{-3} \text{ Mpc}^{-3})$		$(10^{11} h^{-1} M_{\odot})$
0.24	0.37	−18.5	−18.0	89.6	88.6	7.51	0.273	2.43
4.626		−19.0	−18.5	92.4	91.9	6.22	0.259	3.58
		−19.5	−19.0	94.6	93.5	4.89	0.221	4.58
		−20.0	−19.5	95.4	94.7	3.41	0.153	5.48
		−20.5	−20.0	96.2	95.4	1.8	0.1	6.22
		None	−20.5	97.3	96.2	1.02	0.073	8.66
0.37	0.5	−18.5	−18.0	81.9	81.7	8.14	0.291	2.34
8.262		−19.0	−18.5	87.5	86.9	6.73	0.275	3.53
		−19.5	−19.0	90.8	90.8	5.44	0.242	4.6
		−20.0	−19.5	93.2	92.9	3.88	0.179	5.36
		−20.5	−20.0	94.1	94.2	2.1	0.113	6.05
		None	−20.5	96.1	96.1	1.22	0.079	8.79
0.5	0.63	−19.0	−18.5	81.2	82.0	6.22	0.273	3.31
12.22		−19.5	−19.0	87.0	86.2	5.31	0.234	4.28
		−20.0	−19.5	90.2	89.9	3.91	0.174	5.03
		−20.5	−20.0	92.2	91.7	2.15	0.117	5.78
		None	−20.5	95.2	95.0	1.25	0.076	8.26
0.63	0.76	−19.5	−19.0	82.2	83.7	4.96	0.232	4.09
16.177		−20.0	−19.5	86.9	86.7	3.91	0.177	4.85
		−20.5	−20.0	89.9	89.6	2.23	0.118	5.5
		None	−20.5	94.1	93.9	1.3	0.08	8.08
0.76	0.89	−20.5	−20.0	87.9	87.4	2.62	0.129	5.42
19.922		None	−20.5	93.1	92.9	1.53	0.083	7.99

Table B2. Table of galaxy clustering samples used in this analysis including colour splits. Completeness, purity, and satellite fraction are defined in the text. \bar{n} is the number density of the sample.

z-min	z-max	Colour	M_B^h bright	M_B^h faint	Comp (%)	Purity (%)	\bar{n}	Sat frac	Median M_{halo}
Volume	$(10^6 h^{-3} \text{Mpc}^{-3})$						$(10^{-3} h^{-3} \text{Mpc}^{-3})$		$(10^{11} h^{-1} M_{\odot})$
0.24	0.37	Red	−18.5	−18.0	73.9	63.9	3.43	0.481	13.3
4.626			−19.0	−18.5	85.8	80.2	3.11	0.47	19.4
			−19.5	−19.0	92.2	88.2	2.52	0.401	20.4
			−20.0	−19.5	93.5	91.6	1.68	0.286	20.9
			−20.5	−20.0	94.9	92.8	0.788	0.193	23.2
			None	−20.5	96.6	94.6	0.454	0.13	64.7
		Blue	−18.5	−18.0	71.7	78.3	4.08	0.097	1.91
			−19.0	−18.5	81.8	86.2	3.11	0.048	2.36
			−19.5	−19.0	88.9	91.0	2.37	0.029	2.89
			−20.0	−19.5	92.2	92.6	1.73	0.023	3.63
			−20.5	−20.0	94.0	94.3	1.01	0.026	4.69
			None	−20.5	95.9	95.5	0.571	0.028	6.37
0.37	0.5	Red	−18.5	−18.0	56.3	46.3	3.94	0.412	6.61
8.262			−19.0	−18.5	69.5	63.6	3.4	0.426	12.2
			−19.5	−19.0	81.1	76.9	2.88	0.401	16.7
			−20.0	−19.5	88.2	84.0	2	0.31	17.9
			−20.5	−20.0	91.2	88.5	0.964	0.212	22.2
			None	−20.5	94.1	93.9	0.571	0.136	50.7
		Blue	−18.5	−18.0	55.2	64.2	4.2	0.178	1.95
			−19.0	−18.5	66.3	71.1	3.33	0.12	2.4
			−19.5	−19.0	76.3	81.0	2.55	0.062	2.88
			−20.0	−19.5	84.5	88.3	1.88	0.04	3.51
			−20.5	−20.0	89.3	91.7	1.14	0.029	4.41
			None	−20.5	94.2	94.4	0.651	0.028	6.02
0.5	0.63	Red	−19.0	−18.5	64.2	61.0	2.94	0.448	12.7
12.22			−19.5	−19.0	74.9	73.5	2.58	0.4	14.9
			−20.0	−19.5	83.9	82.8	1.86	0.316	16.6
			−20.5	−20.0	88.8	87.7	0.94	0.219	18.1
			None	−20.5	93.4	93.5	0.579	0.129	36.4
		Blue	−19.0	−18.5	65.2	69.5	3.27	0.114	2.29
			−19.5	−19.0	75.6	75.7	2.73	0.076	2.8
			−20.0	−19.5	84.1	84.8	2.05	0.044	3.35
			−20.5	−20.0	89.5	89.5	1.21	0.038	4.17
			None	−20.5	94.1	93.7	0.67	0.03	5.78
0.63	0.76	Red	−19.5	−19.0	66.5	65.2	2.42	0.378	11.6
16.177			−20.0	−19.5	76.1	73.6	1.88	0.302	12.9
			−20.5	−20.0	83.4	81.4	0.989	0.214	14.8
			None	−20.5	91.3	91.3	0.61	0.132	27.7
		Blue	−19.5	−19.0	67.0	70.6	2.54	0.093	2.77
			−20.0	−19.5	76.2	78.3	2.02	0.06	3.32
			−20.5	−20.0	84.4	85.5	1.25	0.041	4.01
			None	−20.5	91.7	91.5	0.686	0.034	5.56
0.76	0.89	Red	−20.5	−20.0	77.8	74.7	1.22	0.214	12.1
19.922			None	−20.5	88.7	87.9	0.735	0.129	22.6
		Blue	−20.5	−20.0	78.5	80.4	1.41	0.054	3.88
			None	−20.5	88.5	88.9	0.795	0.041	5.39

with redshift. These trends are also seen in fainter samples but aren't as clear once errors are included. Brighter galaxies live in larger haloes, and this trend is particularly strong for red galaxies. These red galaxies also on average live in significantly larger haloes

than their blue counterparts with the same luminosity and redshift. At fixed colour and luminosity, the median halo mass increases with decreasing redshift as the dark matter growth rate is large on small non-linear scales over this redshift range.

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